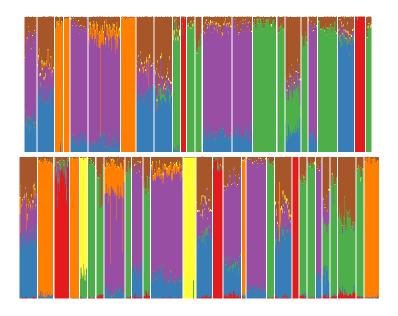
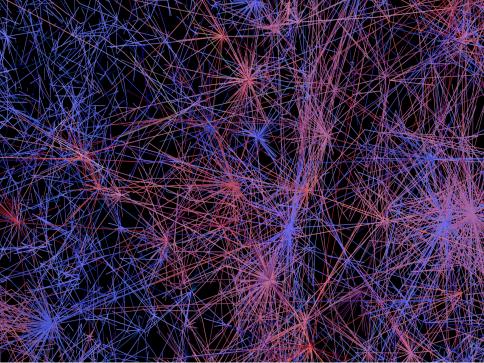
MODERN PROBABILISTIC MODELING FOR MASSIVE DATA

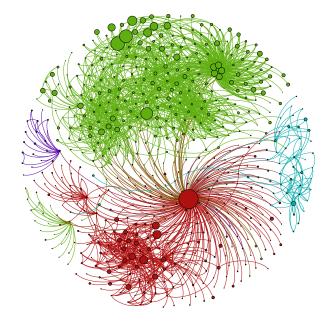
David M. Blei Columbia University

Modern probabilistic modeling:	
An efficient framework for discovering useful patterns in massive data.	



Population analysis of 2 billion genetic measurements



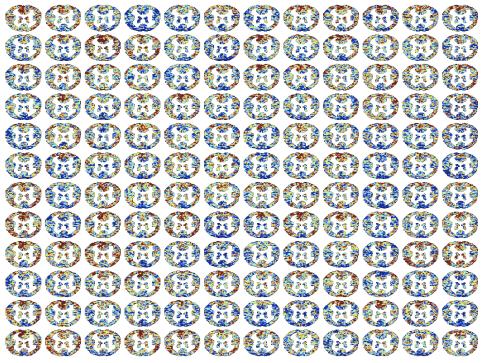


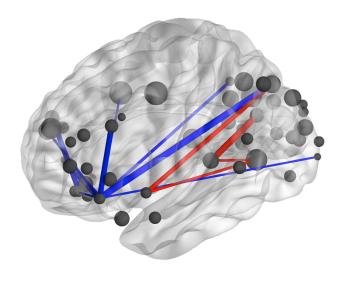
Communities discovered in a 3.7M node network of U.S. Patents



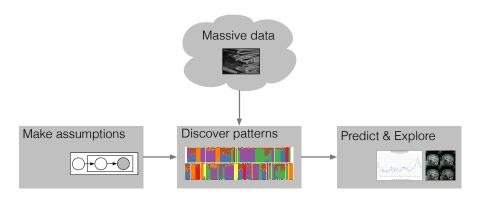


Topics found in 1.8M articles from the New York Times



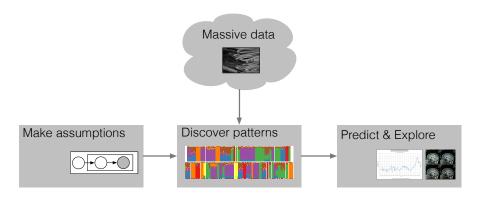


Neuroscience analysis of 220 million fMRI measurements



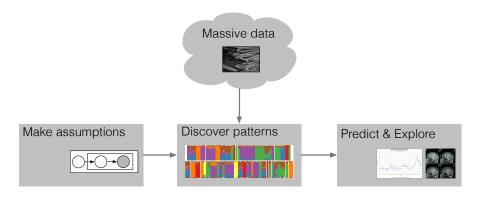
Our perspective:

- ▶ This is a framework for **customized data analysis**, crucial to many fields.
- ▶ The pipeline separates assumptions, computation, application
- It facilitates solving modern data science problems.



Our goal:

Develop modeling into a flexible, powerful and easy-to-use way to solve real-world problems.



Our challenges:

- Develop new ways to build flexible models
- Develop algorithms that work on many problems and with massive data.
- Solve new problems in science, industry, and government



Jaan Aaltosar

Allison Chaney

Rajesh Ranganath

Maja Rudolph



Laurent Charlin

Alp Kucukelbir

Stephan Mandt

Jeremy Manning

James McInerney

Probabilistic topic models Powerful and flexible algorithms for analyzing massive collections of text

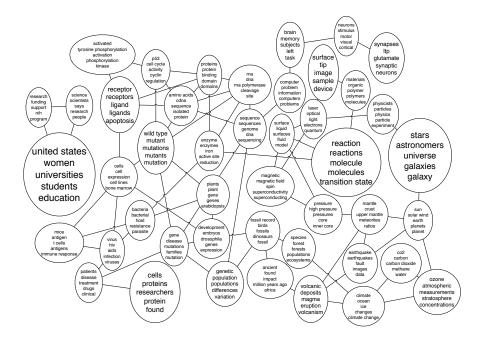


- ▶ ORGANIZE
- **▶ VISUALIZE**
- **► SUMMARIZE**
- ► SEARCH
- ► PREDICT
- **▶ UNDERSTAND**



TOPIC MODELING

- Discover the thematic structure
- 2. Annotate the documents
- 3. **Use** the annotations to visualize, organize, summarize, ...



Seeking Life's Bare (Genetic) Necessities

COLD SPRING HARBOR, NEW YORK-How many genes does an organism need to survive? Last week at the genome meeting here,* two genome researchers with radically different approaches presented complementary views of the basic genes needed for life. One research team, using computer analyses to compare known genomes, concluded that today's organisms can be sustained with just 250 genes, and that the earliest life forms required a mere 128 genes. The other researcher mapped genes in a simple parasite and estimated that for this organism.

job-but that anything short of 100 wouldn't be enough. Although the numbers don't match precisely, those predictions

800 genes are plenty to do the

Genes Mycoplasma genome 469 genes * Genome Mapping and Sequenc-

genome 1703 genes

"are not all that far apart," especially in comparison to the 75,000 genes in the human genome, notes Siv Andersson of Uppsala University in Sweden, who arrived at the 800 number. But coming up with a consensus answer may be more than just a genetic numbers game, particularly as more and more genomes are completely mapped and sequenced. "It may be a way of organizing any newly sequenced genome," explains

Arcady Mushegian, a computational molecular biologist at the National Center for Biotechnology Information (NCBI) in Bethesda, Maryland. Comparing an

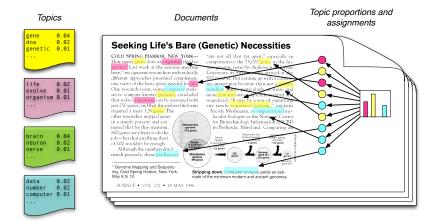
Redundant and needed -122 genes nathways +22 genes Minimal 250 genes Ancestral

Stripping down. Computer analysis yields an estimate of the minimum modern and ancient genomes.

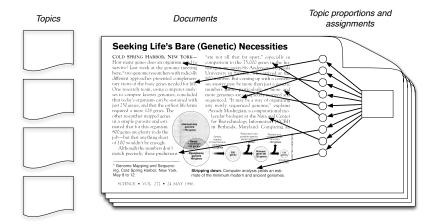
ing, Cold Spring Harbor, New York, May 8 to 12.

SCIENCE • VOL. 272 • 24 MAY 1996

Documents exhibit multiple topics.



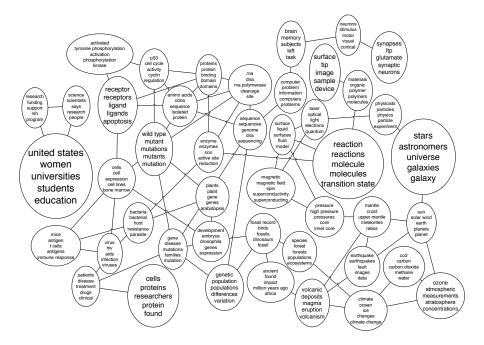
Latent Dirichlet Allocation



Latent Dirichlet Allocation

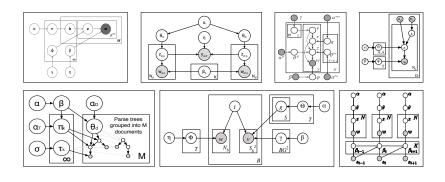


- ▶ **Data**: The OCR'ed collection of *Science* from 1990–2000
 - 17K documents
 - 11M words
 - 20K unique terms (stop words and rare words removed)
- Model: 100-topic LDA model using variational inference.

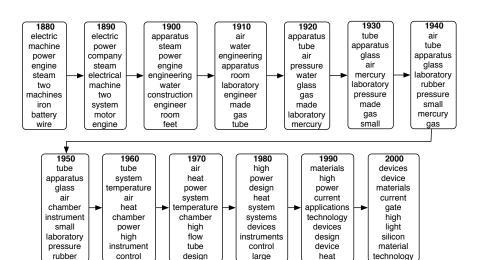




- ▶ LDA builds on decades of research about how to **derive meaning from text**.
- LDA more easily scales to massive data and generalizes to new data.
- LDA has had a big impact on many fields
 - Natural language processing
 - Computer vision
 - Recommendation systems
 - Web search
 - Computational biology and genetics



- ▶ LDA is a simple **building block** that enables many applications.
- ▶ Each model solves a different problem, fuses different kinds of data.
- Models and their algorithms easily compose.





SKY WATER TREE MOUNTAIN PEOPLE



SCOTLAND WATER FLOWER HILLS TREE



SKY WATER BUILDING PEOPLE WATER



FISH WATER OCEAN TREE CORAL



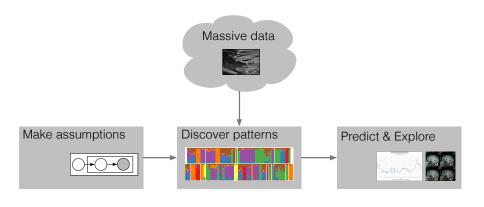
PEOPLE MARKET PATTERN
TEXTILE DISPLAY



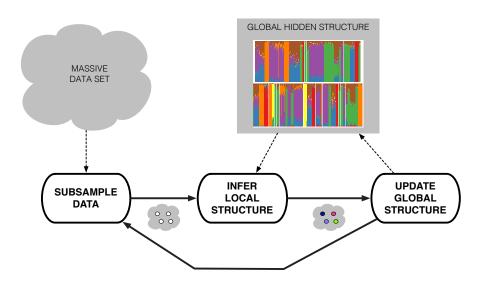
BIRDS NEST TREE BRANCH LEAVES



Probabilistic inference Given a model, use an algorithm to discover the hidden patterns in the data.



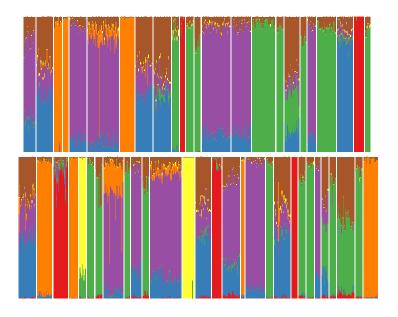
- ▶ Probabilistic inference is the main algorithmic & statistical problem.
- We square the modeling assumptions with the observed data. E.g., which topics likely generated a collection of documents?
- We need scalable and generic inference.



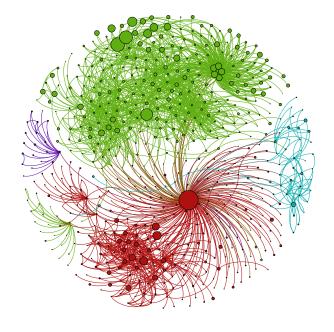
Stochastic variational inference scales to massive data.



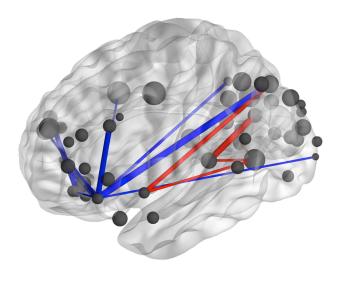
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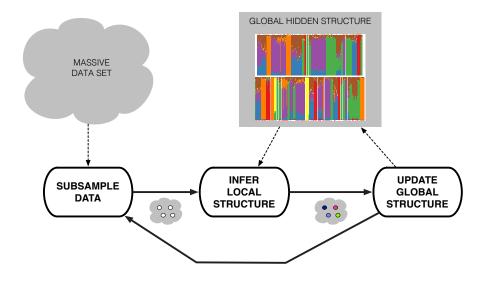
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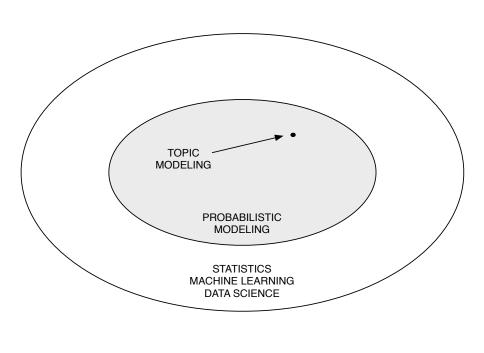


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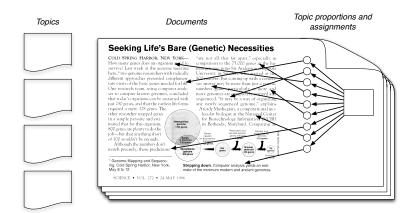


- ▶ Uses stochastic optimization (Robbins & Monro, 1951)
- Scales up 50 years of research in Bayesian modeling
- Though these are recent results, they have been adapted to many domains

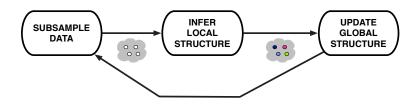
Modern probabilistic modeling:	
An efficient framework for discovering useful patterns in massive data.	



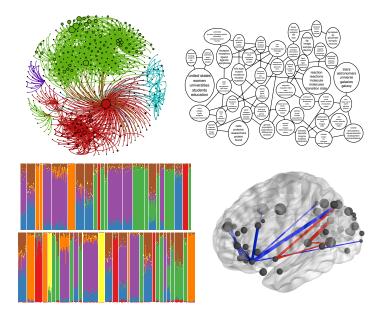
I. Assume our data come from a model with hidden patterns at work

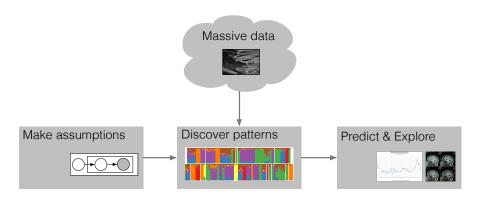


II. Discover those patterns in the data



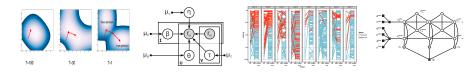
III. Use the discovered patterns to predict about and explore the data





Our goal:

Develop modeling into a flexible, powerful and easy-to-use way to solve real-world problems.



Models and applications

- genetic measurements
- hierarchical topics in a corpus
- changing preferences
- equations and text
- newsworthy events in Twitter

- news consumption in a network
- word meanings
- counselor/patient dialogs
- declassified cables from the 70s
- neural readings in a fish

Inference

- active subsampling
- averaged gradients
- annealing and inference

- stochastic optimization
- structured variational inference
- probabilistic programming



We should seek out unfamiliar summaries of observational material, and establish their useful properties... And still more novelty can come from finding, and evading, still deeper lying constraints.

(John Tukey, The Future of Data Analysis, 1962)